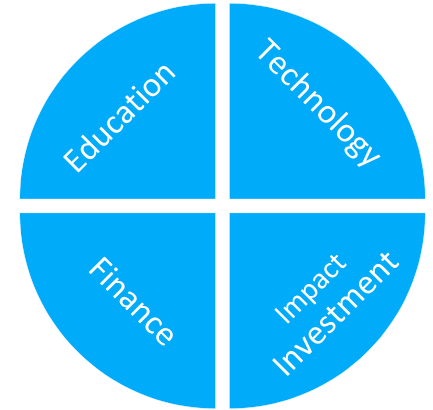


“

“We should be designing a society that, doesn’t leave behind someone who accidentally got the wrong ball (lottery of birth)”.

– Mr. Warren Buffet



“

“Education is not just a need, but a tool to alleviate poverty”

Our MVP



Meet: Amit & Dave



AMIT

Born: India

Undergraduate: IIT,
Delhi

Work Exp.: 4 Yrs

GMAT Score: 750

Graduate Study:
Cornell MBA

Funding Need: \$80K

Interest Rate: **13.5%**



DAVE

Born: USA

Undergraduate: UofR,
Rochester NY

Work Exp.: 4 Yrs

GMAT Score: 750

Graduate Study: Cornell
MBA

Funding Need: \$80K

Interest Rate: **6.5%**

Human Capital with a drive to thrive in life is an asset worth investing.

Project Scope:

To predict importance of education-related variables for loan default

- ☐ Finding co-relationship between the education level and credit worthiness of an individual
- ☐ We will use [Lending Club loan dataset](#) to train our model

Questions to be addressed

1. Is there any close correlation between/among the **educational parameters** in predicting someone's **credit worthiness** given the training dataset.
2. Educational parameters could be
 - a) education level
 - b) studied major
 - c) post graduation experience
 - d) college attended etc.
3. Credit worthiness dimensions could be
 - a) how much loan can be approved
 - b) at what interest rates
 - c) what loan terms (in years)
 - d) what is the probability the loan can go default

Dataset

Notes:

1. Dataset has 42K+ records with ~130 attributes
2. Employer Title replaces Employer Name for all loans listed after 9/23/2013
3. Some fields can be directly considered and some to be calculated/inferred

Following are some example fields: In the data dictionary file, I have tried to highlight the factors column that I think will be useful for our study. As we progress, we can add or delete more fields.

Factors	LoanStatNew	Description
	delinq_amnt	The past-due amount owed for the accounts on which the borrower is now delinquent.
	desc	Loan description provided by the borrower
	dti	A ratio calculated using the borrower's total monthly debt payments on the total debt obligations, excluding mortgage and the requested LC loan, divided by the borrow
	dti_joint	A ratio calculated using the co-borrowers' total monthly payments on the total debt obligations, excluding mortgages and the requested LC loan, divided by the co-borr
	earliest_cr_line	The month the borrower's earliest reported credit line was opened
	emp_length	Employment length in years. Possible values are between 0 and 10 where 0 means less than one year and 10 means ten or more years.
	emp_title	The job title supplied by the Borrower when applying for the loan.*
	fico_range_high	The upper boundary range the borrower's FICO at loan origination belongs to.
	fico_range_low	The lower boundary range the borrower's FICO at loan origination belongs to.
	funded_amnt	The total amount committed to that loan at that point in time.
	funded_amnt_inv	The total amount committed by investors for that loan at that point in time.
	grade	LC assigned loan grade
	home_ownership	The home ownership status provided by the borrower during registration or obtained from the credit report. Our values are: RENT, OWN, MORTGAGE, OTHER
	id	A unique LC assigned ID for the loan listing.
	il_util	Ratio of total current balance to high credit/credit limit on all install acct
	initial_list_status	The initial listing status of the loan. Possible values are – W, F
	inq_fi	Number of personal finance inquiries
	inq_last_12m	Number of credit inquiries in past 12 months

Backup Slides

Serviceable Available Market (SAM)

1M+

INTERNATIONAL STUDENTS TO USA
/YEAR (TOTAL)
Total Available Market*

500K+

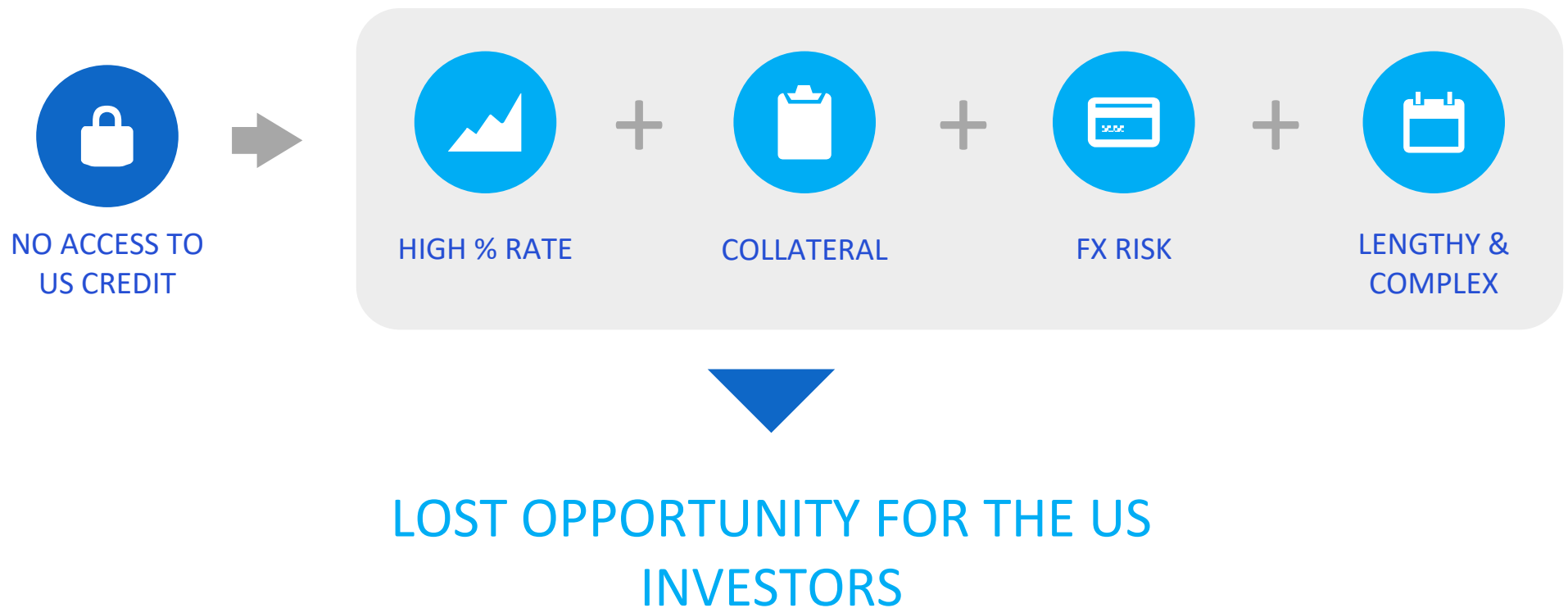
MBA, FINANCE, STEM FIELD
Serviceable Available Market
(52% of the Total*)

\$10B+

AVERAGE LOAN SIZE of
\$20K/Year

**Ref: opendoors data portal*

Problem: inefficient & painful process



Our solution



P2P PLATFORM

- Alumni based lending
- Alternate credit risk model



LOWER % RATE

- We source 'friendly' sources of capital by tapping into alumni networks ("Freeformers")



EASY & FAST

- Online (LinkedIn)
- 5 – 6 mins to fill in an application
- Followed by a 20 mins of phone /skype call



MENTORING & COACHING


- Improves repayment rates
- Borrowers becomes the investors



For any questions /clarifications,
please contact:

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Thank You